## **Overview**

This report was generated using ChIPQC

The report provides both general and ChIP-seq specific quality metrics and diagnostic graphics to allow for the quantitative assessment of ChIP-seq quality.

The report is split into three main sections:

- QC Summary Overview of results.
- QC Results Full QC results and figures.
- QC files and versions Files and program versions used in QC

# **QC Summary**

**Table 1.** Summary of ChIP-seq filtering and quality metrics.

ID	Tissue	Factor	Condition	Replicate	Reads	Dup%	ReadL	FragL	RelCC	SSD	RiP%
DU145MetR_input	DU145WT	input	DU145MetR_input	1	3506491	18	146	293	0.38	2.7	0.83
DU145MetR_RUNX3	DU145MetR	RUNX3	DU145MetR_RUNX3	1	3166603	26	148	297	0.7	5	4
DU145MetR_SRF	DU145MetR	SRF	DU145MetR_SRF	1	3365875	21	147	298	0.62	4.7	3
DU145WT_input	DU145WT	input	DU145WT_input	1	3377433	19	149	299	0.79	2.2	0.52
DU145WT_RUNX3	DU145WT	RUNX3	DU145WT_RUNX3	1	2020340	16	148	297	0.62	3.8	4.9
DU145WT_SRF	DU145WT	SRF	DU145WT_SRF	1	3186310	21	148	300	0.65	4.1	4.9

**Table 1**contains a summary of filtering and quality metrics generated by the ChIPQC package. Further information on these metrics, their associated figures and additional quality measures can be found within the related QC Results subsections.

A short description of Table 1 metrics is provided below:

- ID Unique sample ID.
- Tissue/Factor/Condition Metadata associated to sample.
- Replicate Number of replicate within sample group
- Reads Number of sample reads within analysed chromosomes.
- Dup% Percentage of MapQ filter passing reads marked as duplicates
- **FragLen** Estimated fragment length by cross-coverage method
- **SSD** SSD score (htSeqTools)
- FragLenCC Cross-Coverage score at the fragment length
- RelativeCC Cross-coverage score at the fragment length over Cross-coverage at the read length
- RIP% Percentage of reads wthin peaks
- RIBL% Percentage of reads wthin Blacklist regions

# **QC Results**

### Mapping, Filtering and Duplication rate

This section presents the mapping quality, duplication rate and distribution of reads in known genomic features.

Table 2. Number and percantage of mapped, duplicated and MapQ filter passing reads

ID	Tissue	Factor	Condition	Replicate	Unmapped	Mapped	Pass MapQ Filter and Dup	Total Dup%	Pass MapQ Filter%	Pass MapQ Filter and Dup%
DU145MetR_input	DU145WT	input	DU145MetR_input	1	0	3506491	574479	18	90	18
DU145MetR_RUNX3	DU145MetR	RUNX3	DU145MetR_RUNX3	1	0	3166603	713948	25	88	26
DU145MetR_SRF	DU145MetR	SRF	DU145MetR_SRF	1	0	3365875	618470	20	88	21
DU145WT_input	DU145WT	input	DU145WT_input	1	0	3377433	591639	19	91	19
DU145WT_RUNX3	DU145WT	RUNX3	DU145WT_RUNX3	1	0	2020340	292843	16	88	16
DU145WT_SRF	DU145WT	SRF	DU145WT_SRF	1	0	3186310	582294	20	89	21

**Table 2** shows the absolute number of total, mapped, passing MapQ filter and duplicated reads. The percent of mapped reads passing quality filter and marked as duplicates (Non-Redundant Fraction?) are also included. Description of read filtering and flag metrics:

- Total Dup%-Percentage of all mapped reads which are marked as duplicates.
- Pass MapQ Filter%-Percentage of all mapped reads which pass MapQ quality filter
- Pass MapQ Filter and Dup%-Percentage of all reads which pass MapQ filter and are marked asduplicates.

Duplication rates (Dup %) are dependent on the ChIP library complexity and the number of reads sequenced Higher duplication rates maybe due to low ChIP efficiency when read counts are lower or conversely saturation of ChIP signal when sequencing large number of reads. Since this metric is dependent on both read depth and the properties of the ChIP itself, comparison between biological or technical replicates of similat total read counts can best identify problematic libraries .

Highly mappable (multimappable) positions within the genome can attract large levels of duplication and so assessment of duplication before and after MapQ quality filtering can identify contribution of these positions to the duplication rate.



Figure 1. Heatmap of log2 enrichment of reads in genomic features

The distribution of reads across known genomic features such as genes and their subcomponents may allow further evaluation of ChIP-seq success and quality. A transcription factor know to preferentially bind at a genomic feature should show relative enrichment against other transcription factors showing no such preference. In addition, a replicate showing a differing enrichment patterns across genomic features compared to those within its sample group would highlight a potential outlier sample worthy of further investigation

**Figure 2** shows the log2 enrichment of specified genomic features within samples with regions of greater enrichment showing bright yellow and lower enrichment seen in black

#### ChIP signal Distribution and Structure

In this section, metrics relating to genome wide depths of coverage and, the relationship between Watson and Crick reads are presented. The metrics are the SSD metric and cross-coverage metrics, Relative\_CC and fragmentLength\_CC.

Figure 2. Plot of the log2 base pairs of genome at differing read depths



**SSD** is the standard deviation of coverage normalised to the total number of reads. Evaluation of the number of bases at differing read depths,(**figure 3**) alongside the use of the SSD metric allow for an assessment of the distribution of ChIP-seq or input signal.

Successfull Histone and transcription factor ChIP-seq samples will show a higher proportion of genomic positions at greater depths and equivalence of sample and input SSD scores highlights either an unsuccessful ChIP or high levels of anomalous input signal





An important measure of ChIP successive is the degree to which Watson and Crick reads cluster around the centres of transcription factor bindind sites or epigentic marks.

Transcription factor binding sites identified by ChIP-seq will show two distinct peaks of Watson and Crick strands separated by the fragment length. Here the method of cross-coverage (ChIPseq package) analysis is used to investigate this spatial clustering of Watson and Crick reads.

To investigate this spatial clustering, reads on the positive strand are shifted in 1bp steps and the total proportion genome now covered by both strands combined is assessed. **Figure 4** shows the CCov\_Score (described below)

after successive shifts. The points of highest outside of the read-length exclusion region, 2\* the read length, (marked in grey) is considered the fragment length

Following the methodology first presented for cross-correlation by Encode to calculate the Relative Strand Cross Correlation (NSC) and Normalised Strand Cross Correlation, the Relative Cross Coverage score and Fragment Length Cross Coverage score are calculated.

The calculation of cross-coverage (CCov), Relative CCov and Fragment Length CCov scores are explained below:

- CCov\_Score- 1-(Total covered genome size at strand shift)/(covered genome size with no shift)
- Fragment Length CCov- (CCov of fragment length strand shift)/(Minimum CCov)
- Relative CCov- (CCov of fragment length strand shift)/(CCov of read length strand shift)

## **Peak Profile and ChIP Enrichment**

Following the identification of genome wide enrichment (peak calling), the percentage of ChIP signal within enriched regions, as well the average profile across these regions can be used to further evaluate ChIP quality

Figure 4. Plot of the average signal profile across peaks



**Figure5** represents the mean read depth across and around peaks. By identying the average pattern of enrichment across peaks, differences in both mean peak height and shape may be found. This not only assits in a better characterisation of ChIP enrichment but can aid in the identification of outliers.

Figure 5. Barplot of the percentage number of reads in peaks



**Figure6** shows the total percentage of reads contained within enriched regions or peaks. The higher efficiency ChIP-seq will show a higher percentage of reads in enriched regions/peaks and longer epigenetic marks will often have a higher ranges of efficiencies than punctate marks or transcription factors.

Figure 6. Density plot of the number of reads in peaks



**Figure**7 shows the distribution of reads in all peaks. Evaluation of the distibution can allow for greater characteriation of the variability and range of signal in peaks within a sample and so better characterise the signal across peaks than the RIP score may allow.





**Figure8 and 9** shows the correlation between samples as a heatmap and by principal component analysis. Replicate samples of high quality can be expected to cluster together in the heatmap and be spatially grouped within the PCA plot.

# **Files and Versions**

## **R Version Information**

- Version: 4.2.1
- Version\_String :R version 4.2.1 (2022-06-23)

## **ChIPQC Version Information**

- Version: ChIPQC:3.0.0.20230222
- · Author: Tom Carroll, Wei Liu, Ines de Santiago, Rory Stark
- Maintainer: Quanhu Sheng , Tom Carroll , Rory Stark

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